Analysis of Unicorn Startups

Contents

1 Setup

1.1 Import Packages

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from matplotlib.ticker import FuncFormatter
import seaborn as sns
import re
```

2 Data Preparation

2.1 Load Data

```
pd.set_option('display.max_columns', 50, 'display.width', 200)
```

```
df = pd.read_csv('input/Unicorns_Completed.csv')
```

2.2 Data Cleaning

```
def convert_years_months(s):
    m = re.match(r'(\d+)y?\s?(\d+)m?o?', s)
    return f'{m[1]}y{m[2]}m' if m else s

df['Years to Unicorn'] = df['Years to Unicorn'].apply(convert_years_months)
```

2.3 Prepare data

```
df['Unicorn Date'] = pd.to_datetime(df['Unicorn Date'])
df['Valuation ($B)'] = pd.to_numeric(df['Valuation ($B)'])
```

2.4 Preview data

```
df.head()
```

```
|   | Company   | Valuation ($B) | Total Equity Funding ($) | Unicorn Date | Date Founded
↪  | Years to Unicorn | Industry | Country            | City          | Select Investors |
↪  Years to Unicorn (Months)                        |     |
|---+-----------+----------------+--------------------------+--------------+----------------------+---
| 0 | SpaceX    |          350.0 |               9000000000 | Timestamp    | (2012-12-01
↪  00:00:00) |             2002 | 10y3m    | Enterprise Tech    | United States |
↪  Hawthorne        | Opus Capital, RRE Ventures, Relay Ventures        | 123 |
| 1 | ByteDance |          300.0 |               8000000000 | Timestamp    | (2017-04-07
↪  00:00:00) |             2011 | 6y3m     | Enterprise Tech    | China         | Beijing
↪  | Breyer Capital, Parkway VC, TIME Ventures         |  75 |
| 2 | OpenAI    |          157.0 |              18000000000 | Timestamp    | (2019-07-22
↪  00:00:00) |             2015 | 4y6m     | Industrials        | United States | San
↪  Francisco    | Dynamo VC, Susa Ventures, Founders Fund           |  54 |
| 3 | Ant Group |          150.0 |              19000000000 | Timestamp    | (2017-01-01
↪  00:00:00) |             2014 | 3y       | Financial Services | China         | Hangzhou
↪  | Alibaba Group, CPP Investments, The Carlyle Group |  36 |
| 4 | Stripe    |           70.0 |               9000000000 | Timestamp    | (2014-01-23
↪  00:00:00) |             2009 | 5y       | Consumer & Retail  | United States | San
↪  Francisco    | Sequoia Capital China, ZhenFund, K2 Ventures      |  60 |
```

3 Descriptive Analysis

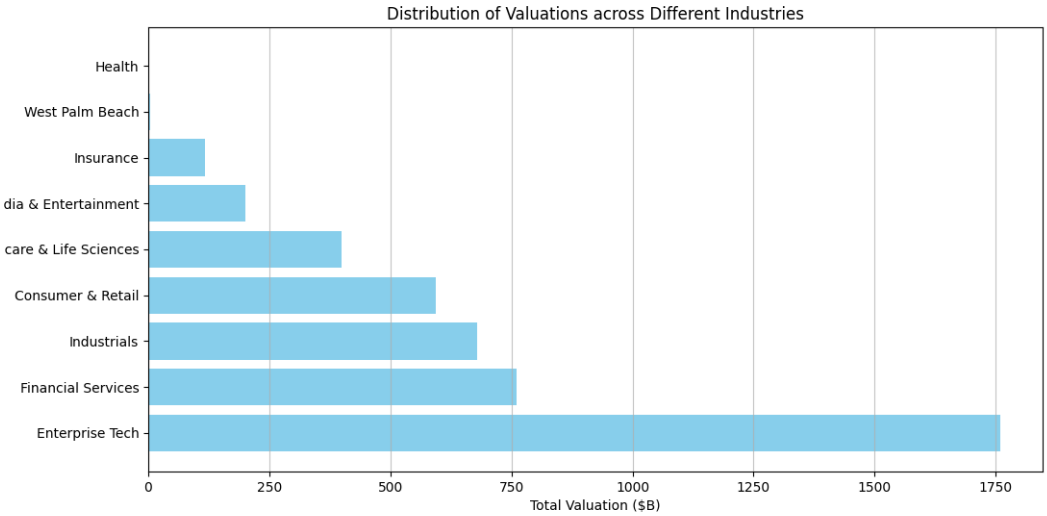
3.1 Distribution of Valuations across Different Industries

```
# Group by industry and sum valuations
```

```
industry_valuation_df = df.groupby('Industry')['Valuation
↪  ($B)'].sum().reset_index().sort_values('Valuation ($B)', ascending=False)
industry_valuation_df
```

	Industry	Valuation ($B)
1	Enterprise Tech	1759.04
2	Financial Services	760.16
5	Industrials	678.55
0	Consumer & Retail	593.3
4	Healthcare & Life Sciences	398.45
7	Media & Entertainment	200.29
6	Insurance	117.06
8	West Palm Beach	3.0
3	Health	1.5

```
plt.figure(figsize=(12, 6))
plt.barh(industry_valuation_df['Industry'], industry_valuation_df['Valuation ($B)'],
↪  color='skyblue')
plt.title('Distribution of Valuations across Different Industries')
plt.xlabel('Total Valuation ($B)')
plt.ylabel('Industry')
plt.grid(axis='x', alpha=0.75)
plt.show()
```



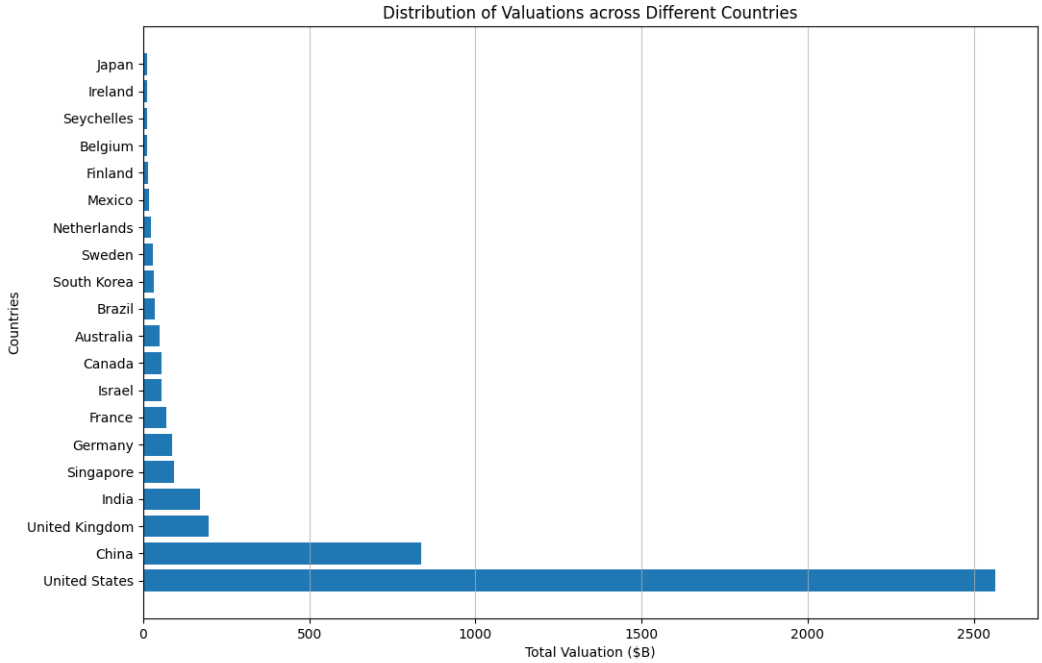
3.2 Distribution of Valuations across Different Countries

```
# Group by Country and sum valuations
country_valuation_df = df.groupby('Country')['Valuation
↪  ($B)'].sum().reset_index().sort_values('Valuation ($B)', ascending=False)[:20]
```

```
country_valuation_df
```

	Country	Valuation ($B)
53	United States	2564.14
10	China	835.65
52	United Kingdom	197.35
24	India	172.07
43	Singapore	92.06
21	Germany	85.9
20	France	70.86
27	Israel	56.22
6	Canada	56.0
1	Australia	48.84
5	Brazil	34.13
45	South Korea	31.34
47	Sweden	29.42
36	Netherlands	24.46
35	Mexico	18.7
19	Finland	14.91
3	Belgium	11.95
42	Seychelles	11.8
26	Ireland	11.05
29	Japan	10.82

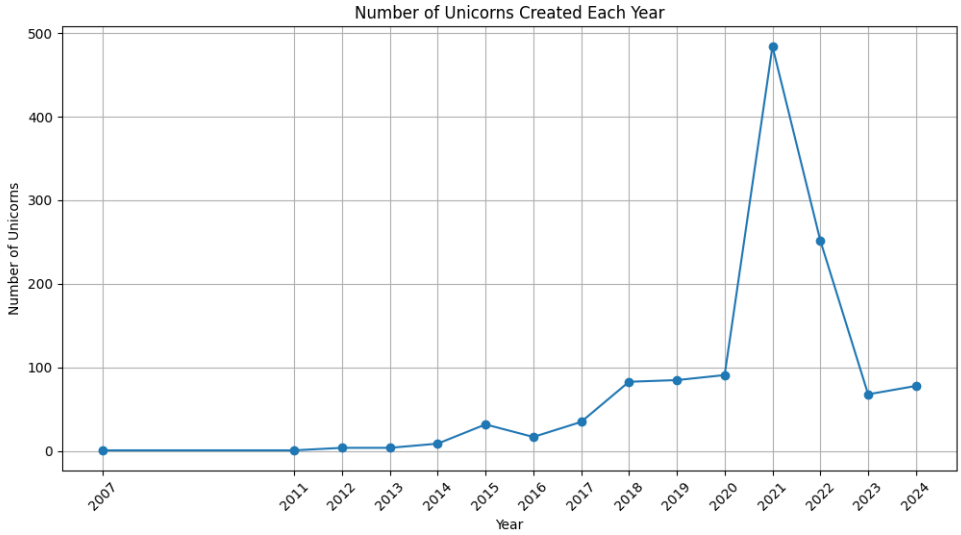
```
plt.figure(figsize=(12, 8))
plt.barh(country_valuation_df['Country'], country_valuation_df['Valuation ($B)'])
plt.title('Distribution of Valuations across Different Countries')
plt.xlabel('Total Valuation ($B)')
plt.ylabel('Countries')
plt.grid(axis='x', alpha=0.75)
plt.show()
```



4 Time-Based Analysis

4.1 Unicorn Growth Over Time

```
unicorn_count = df.groupby(df['Unicorn Date'].dt.year).size()
plt.figure(figsize=(12, 6))
plt.plot(unicorn_count.index, unicorn_count.values, marker='o')
plt.title('Number of Unicorns Created Each Year')
plt.xlabel('Year')
plt.ylabel('Number of Unicorns')
plt.xticks(unicorn_count.index, rotation=45)
plt.grid()
plt.show()
```

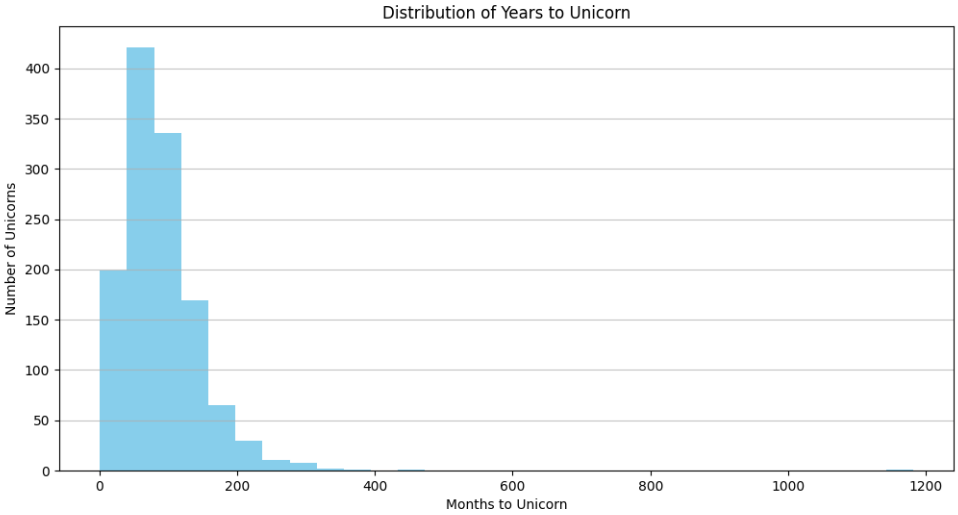


4.2 Years to Unicorn

```
# Function to convert "Years to Unicorn" into total months
def convert_years_to_months(years_str):
    if 'y' in years_str and 'm' in years_str:
        years, months = years_str.split('y')
        months = months.replace('m', '').strip()
        return int(years.strip()) * 12 + int(months)
    elif 'y' in years_str:
        years = years_str.replace('y', '').strip()
        return int(years) * 12
    elif 'm' in years_str:
        months = years_str.replace('mo', '').replace('m', '').strip()
        return int(months)
    else:
        return None

df['Years to Unicorn (Months)'] = df['Years to Unicorn'].apply(convert_years_to_months)
```

```
plt.figure(figsize=(12, 6))
plt.hist(df['Years to Unicorn (Months)'].dropna(), bins=30, color='skyblue')
plt.title('Distribution of Years to Unicorn')
plt.xlabel('Months to Unicorn')
plt.ylabel('Number of Unicorns')
plt.grid(axis='y', alpha=0.75)
plt.show()
```


Distribution of Years to Unicorn
Number of Unicorns
Months to Unicorn
0
50
100
150
200
250
300
350
400
0
200
400
600
800
1000
1200